

Seasonal variations of CH₄ emissions in the Yangtze River Delta region of China are driven by agricultural activities

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THE MAIN CONTENTS



01

Background

03

Results & discussion

02

Material & method

04

Conclusions

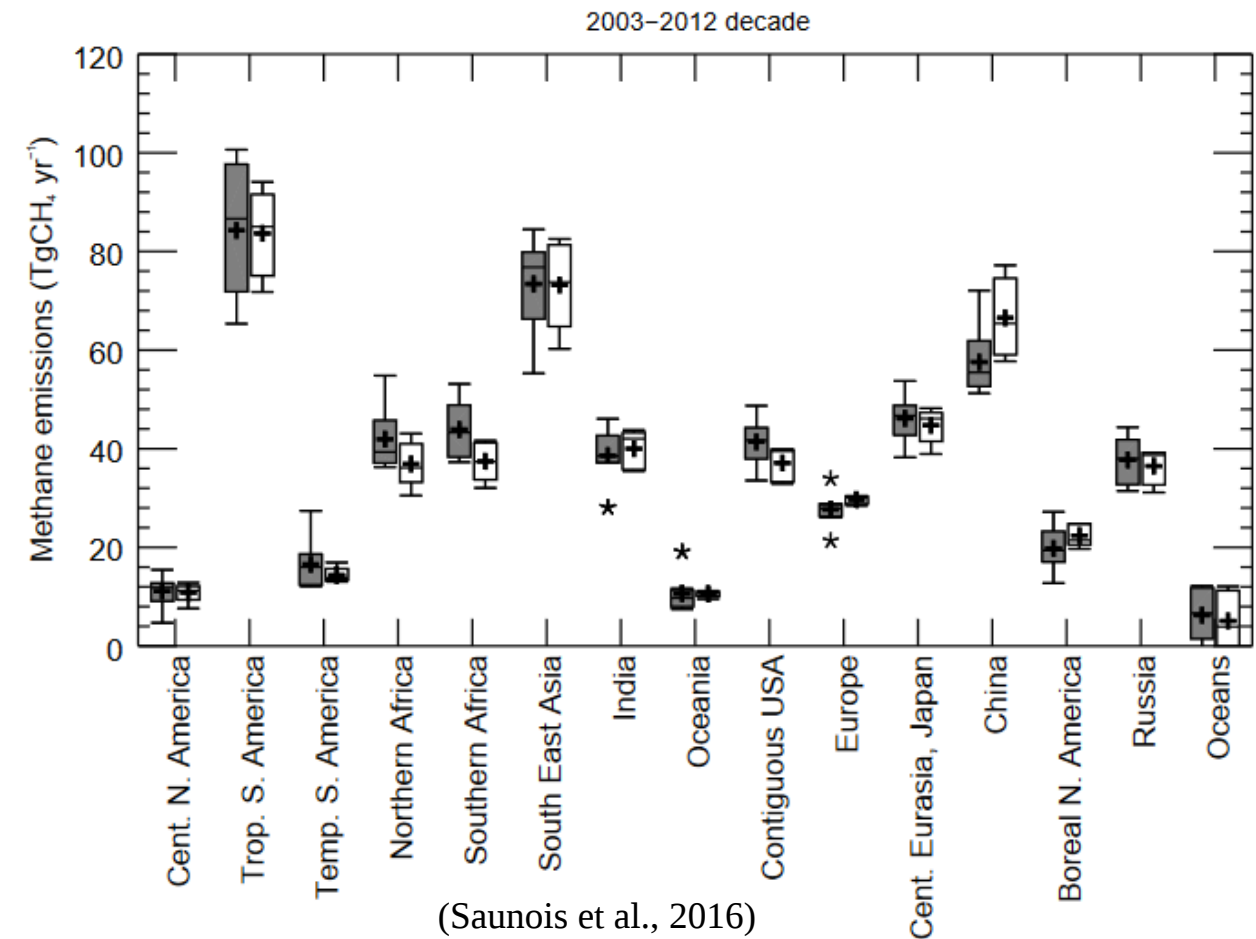


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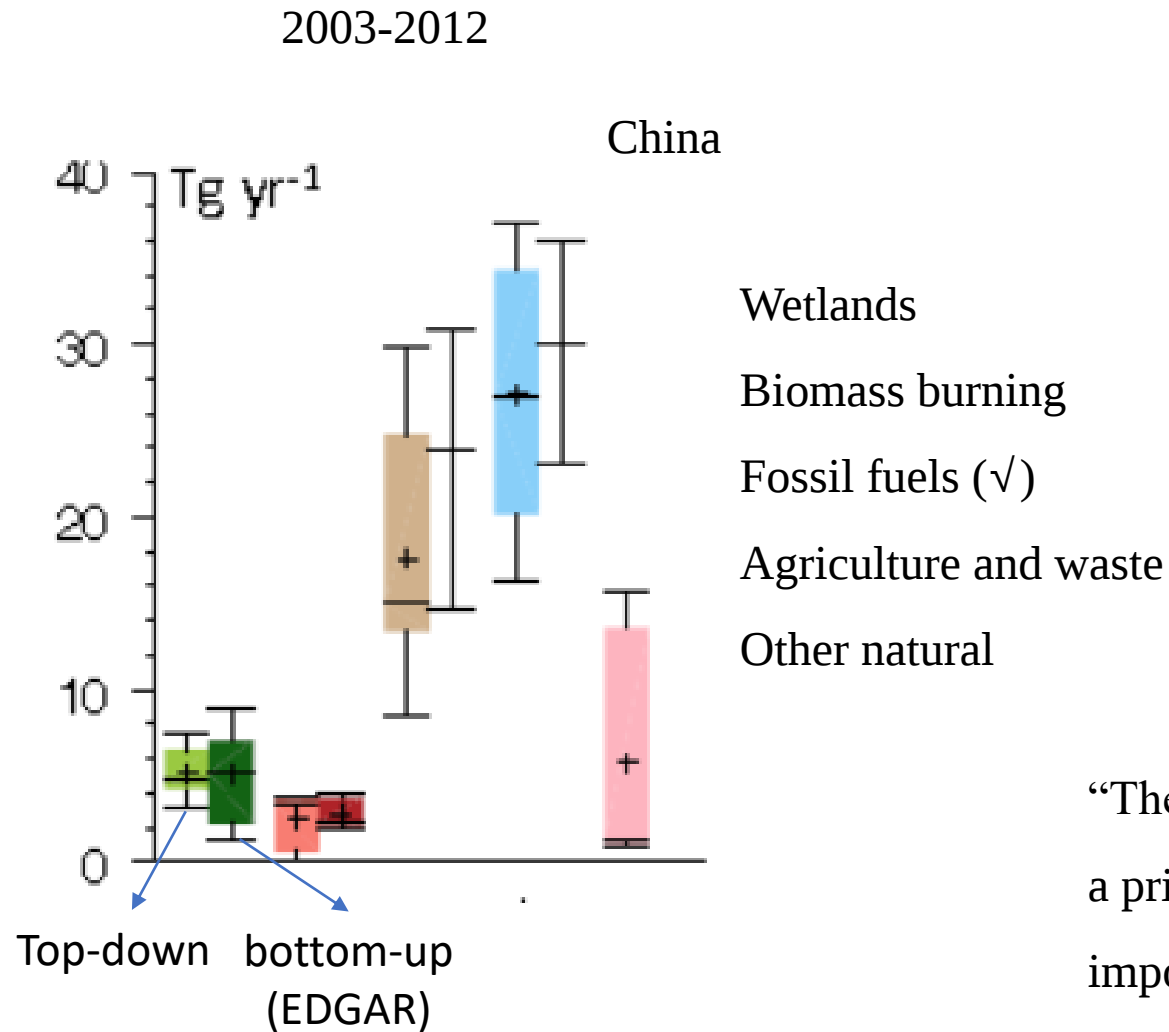
Background

1 Background

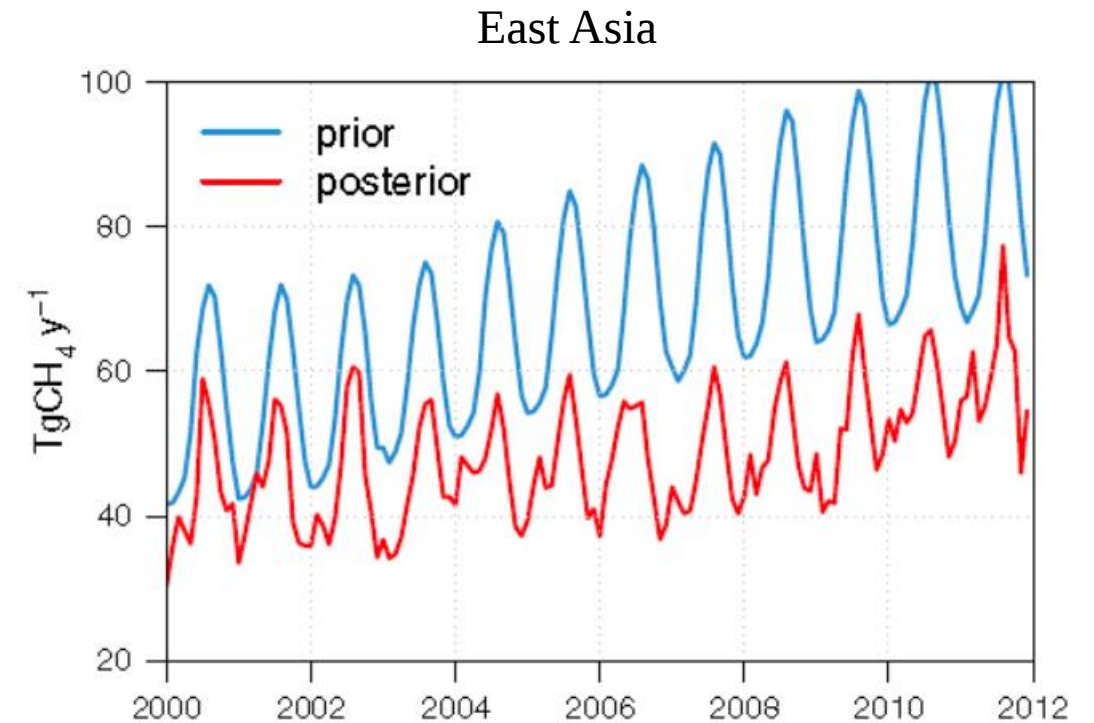
- ✓ Increasing of CH_4 emissions
- ✓ Neglect of natural sources
- ✓ Overestimation of anthropogenic CH_4 emissions in EDGAR



1 Background



(Saunois et al., 2016)



“The greatest difference in emissions, a posteriori with respect to a priori, was found in [eastern China](#), especially in regions important [for rice agriculture](#) and, in particular, [in the summer months](#).”

(Thompson et al. 2015)

Main objectives

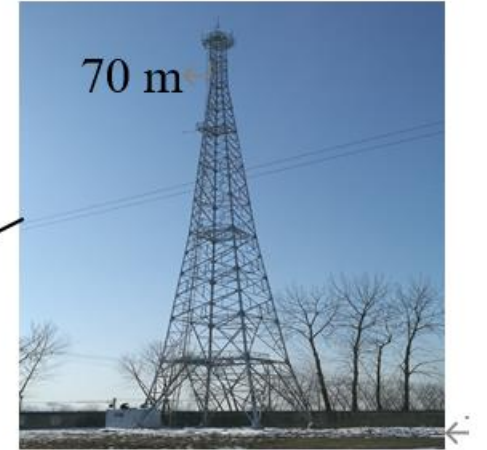
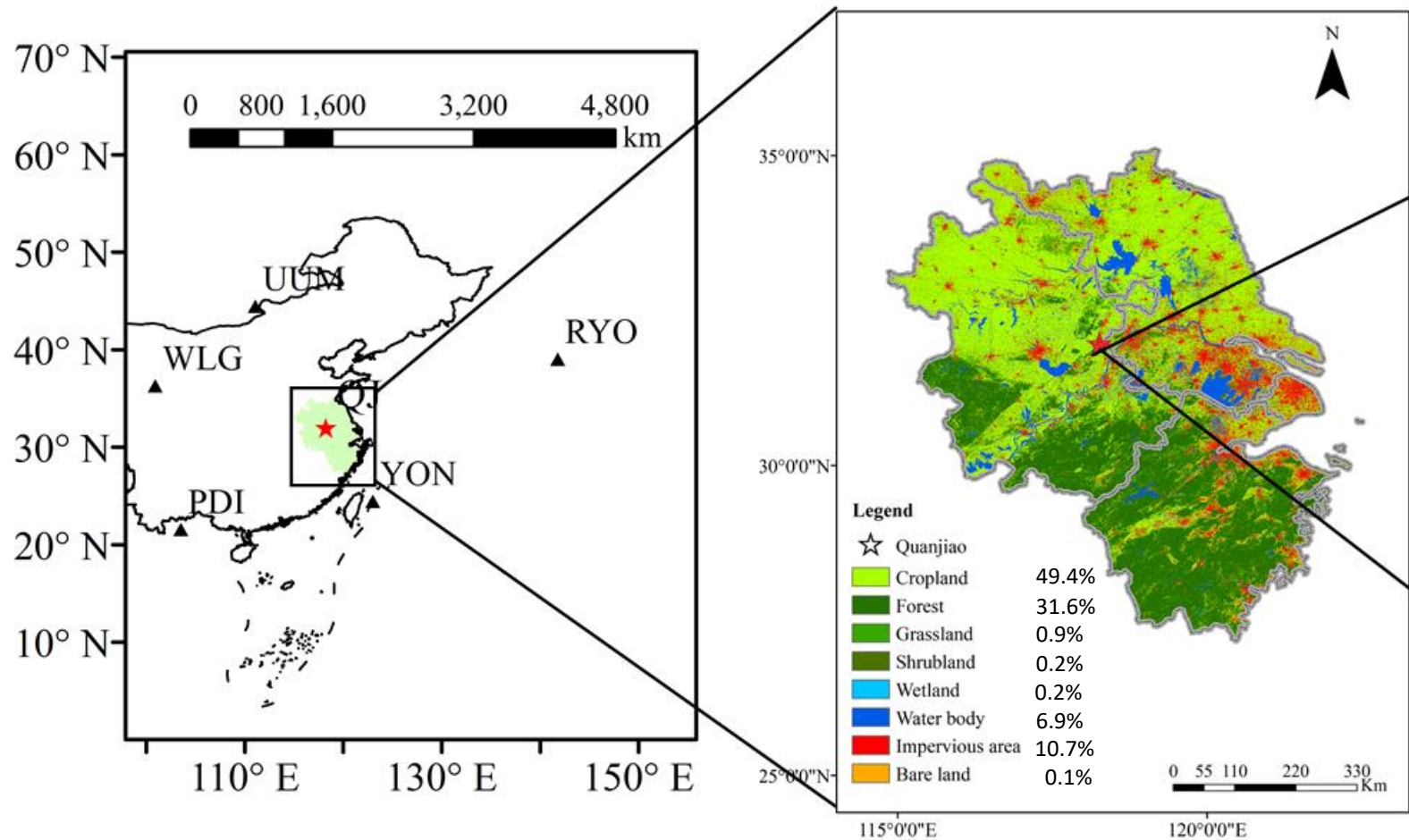
- ❑ Constrain the major CH₄ sources within the Yangtze River Delta (YRD) region at seasonal timescales and identify the main sources that control the seasonality of CH₄ emissions;
- ❑ Assess the relative importance of natural CH₄ emissions from wetlands and water bodies in the YRD region;
- ❑ Evaluate if there are important biases in the *a priori* CH₄ emissions associated with bottom-up inventories (EDGAR);
- ❑ Estimate the total anthropogenic and natural emissions of CH₄ for the YRD region.



2

Material & method

2 Material & method



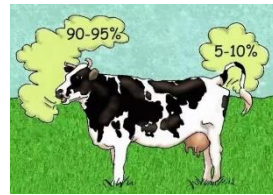
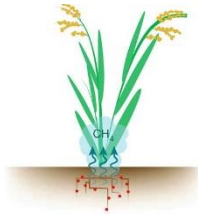
Five WMO/GAW stations:

- (1) Ulaan Uul Mongolia (UUM)
- (2) Waliguang (WLG)
- (3) Pha Din (PDI)
- (4) Yonagunijima (YON)
- (5) Ryori (RYO)

30 m × 30 m
(Gong et al., 2019)

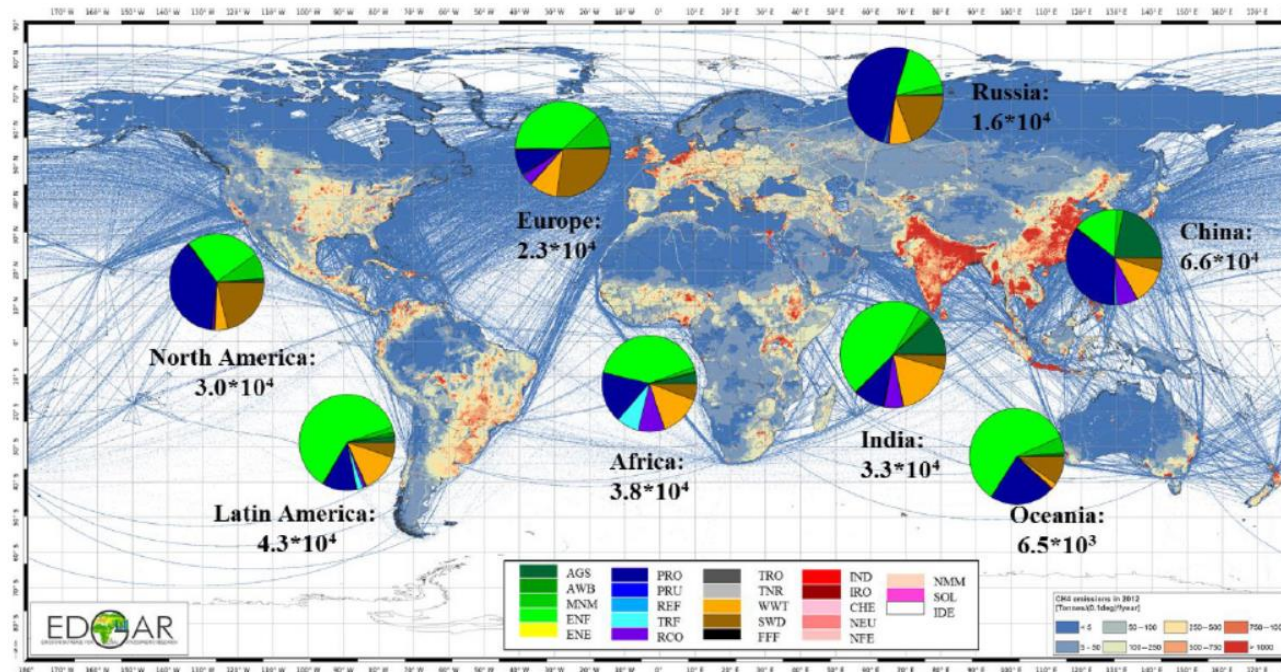
2 Material & method

Anthropogenic
sources



...

EDGAR v432
 $0.1^\circ \times 0.1^\circ$
2012



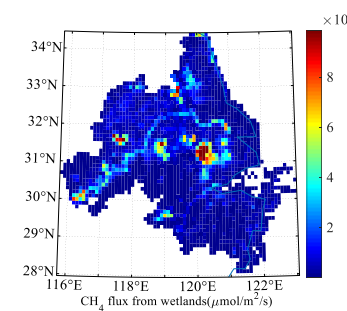
(Janssens-Maenhout et al., 2019)

New version
EDGAR v50

Natural
sources

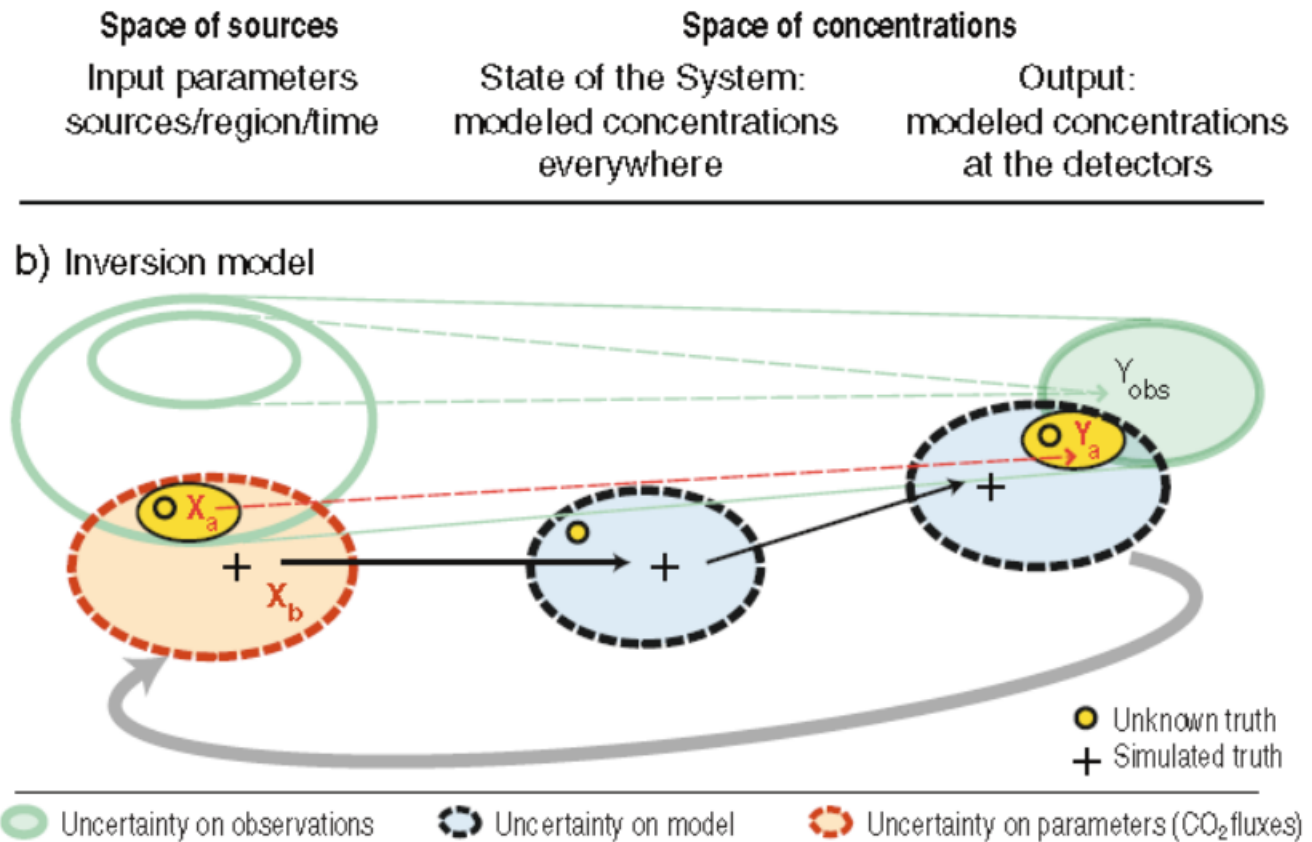


...



$30\text{ m} \times 30\text{ m}$
↓
 $0.1^\circ \times 0.1^\circ$

2 Material & method

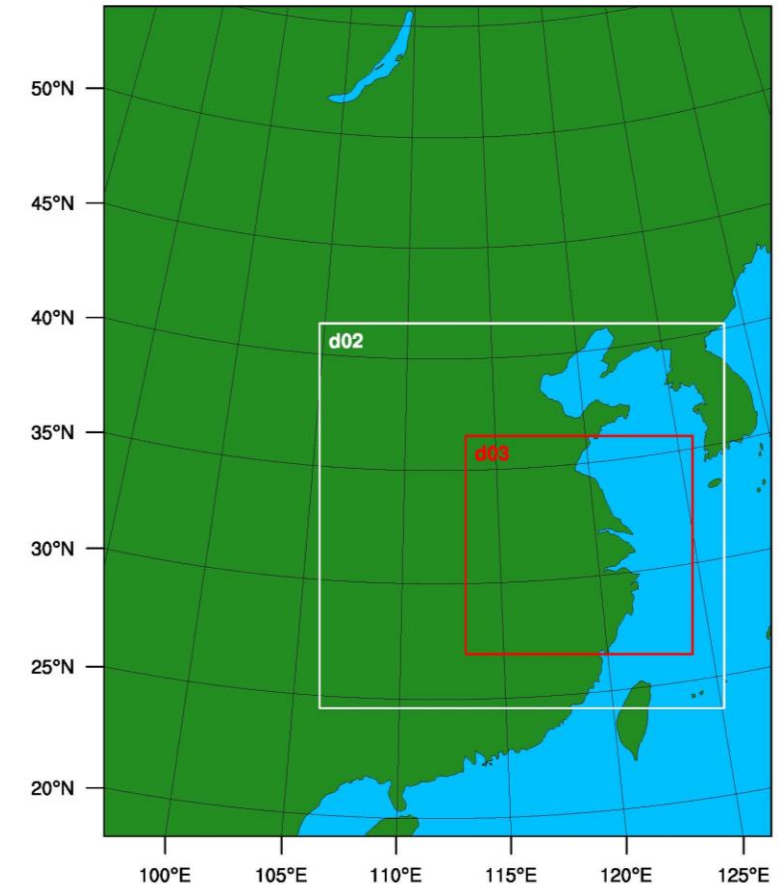


(Ciais et al., 2010)

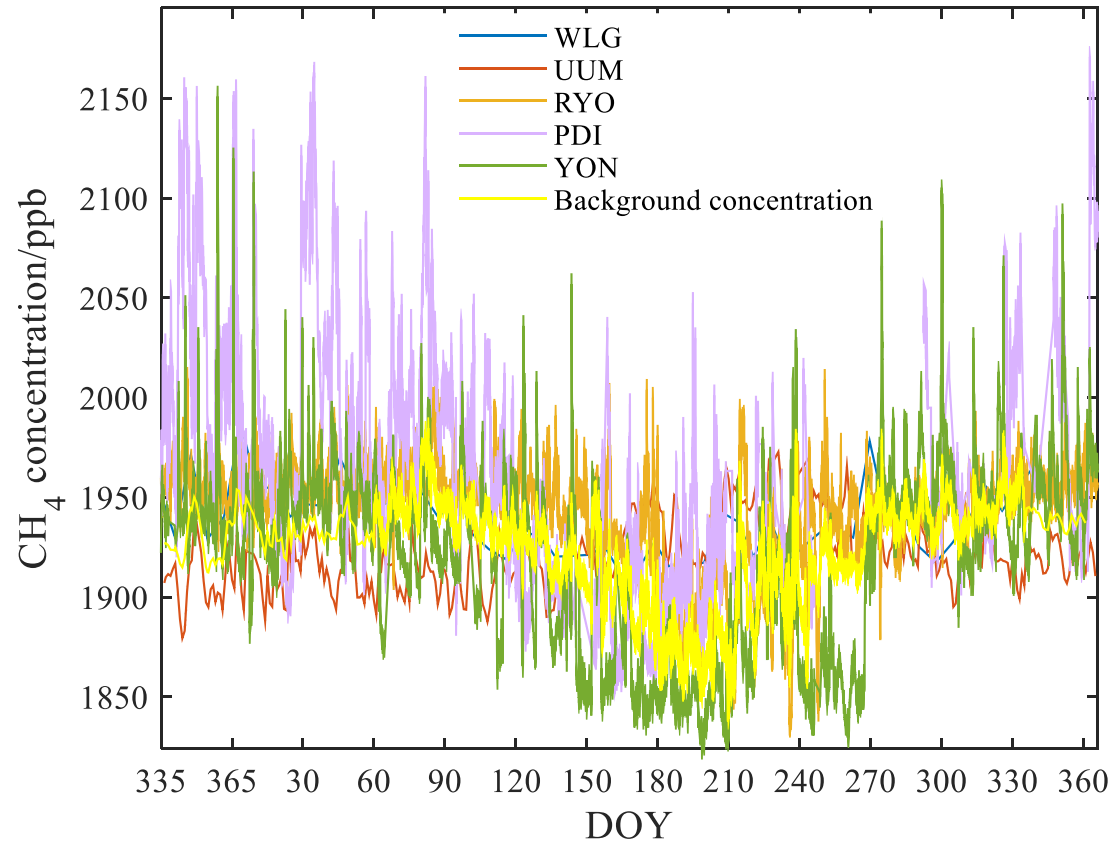
- X_a : posteriori emissions
- X_b : priori emissions
- Y_a : optimized concentration
- Y_{obs} : observed concentration
- ● : true values (assumed)
- + : simulated values

WRF-STILT model

<https://uataq.github.io/stilt/img/animation-example.webm>



The spatial resolutions for d01, d02 and d03 are 27 km, 9 km and 3 km.



A scale factor Bayesian inverse (SFBI) method

Purposes

$$Y_a \quad Y_{\text{obs}}$$

$$\underbrace{\hspace{10em}}_{\approx}$$

Posteriori scaling factors

(?)

A priori scaling factors

(default value is 1)

$$\underbrace{\hspace{15em}}_{\approx}$$

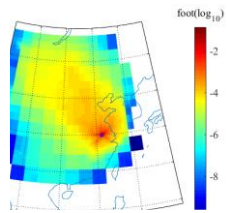


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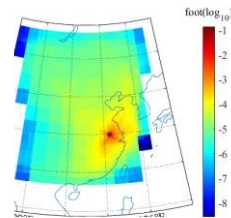
Results & discussion

3 Results & discussion

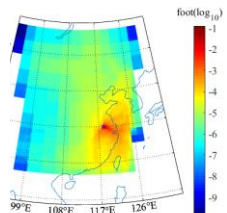
(a)



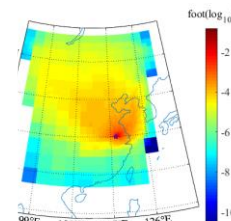
(b)



(c)

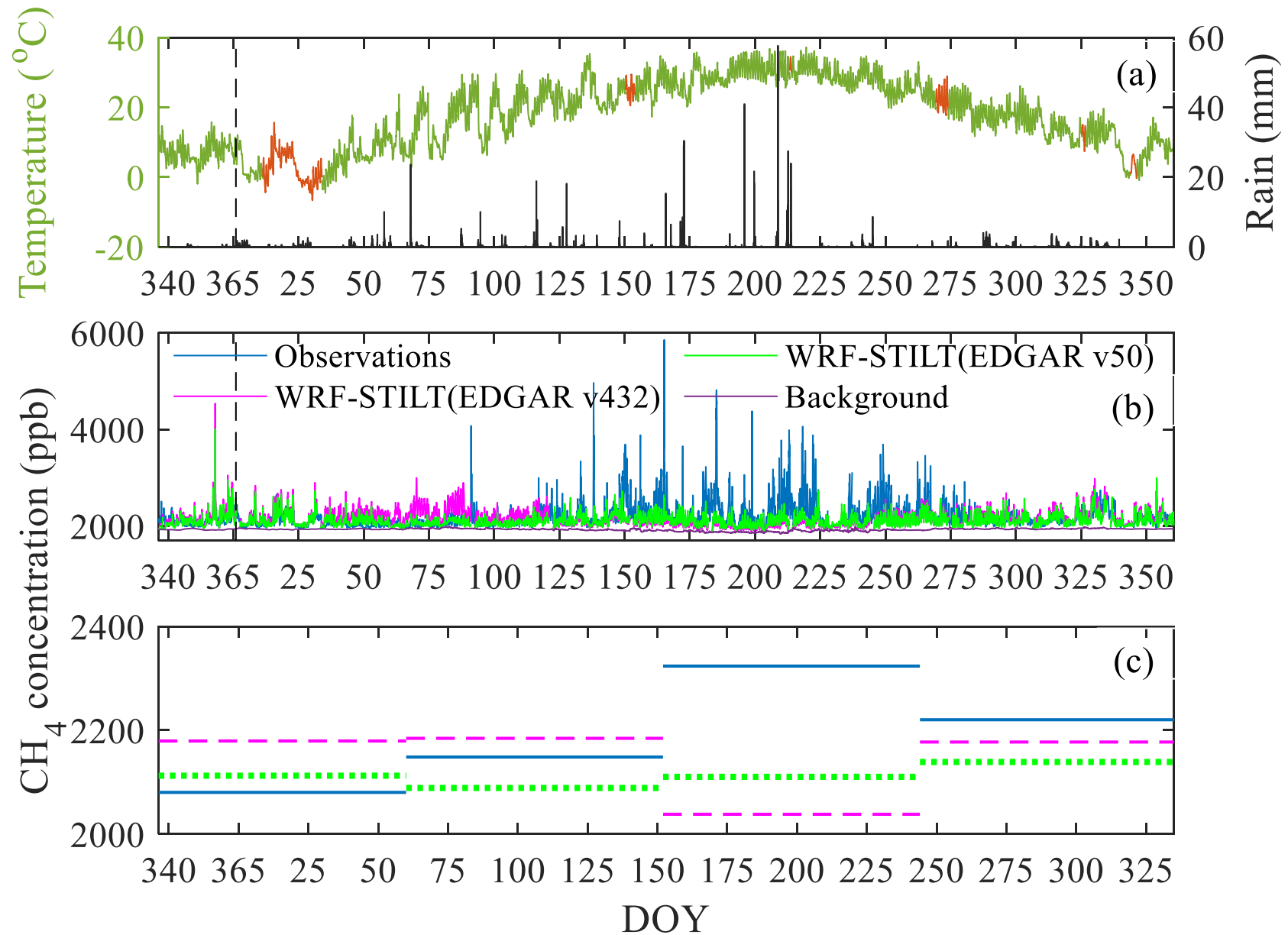


(d)



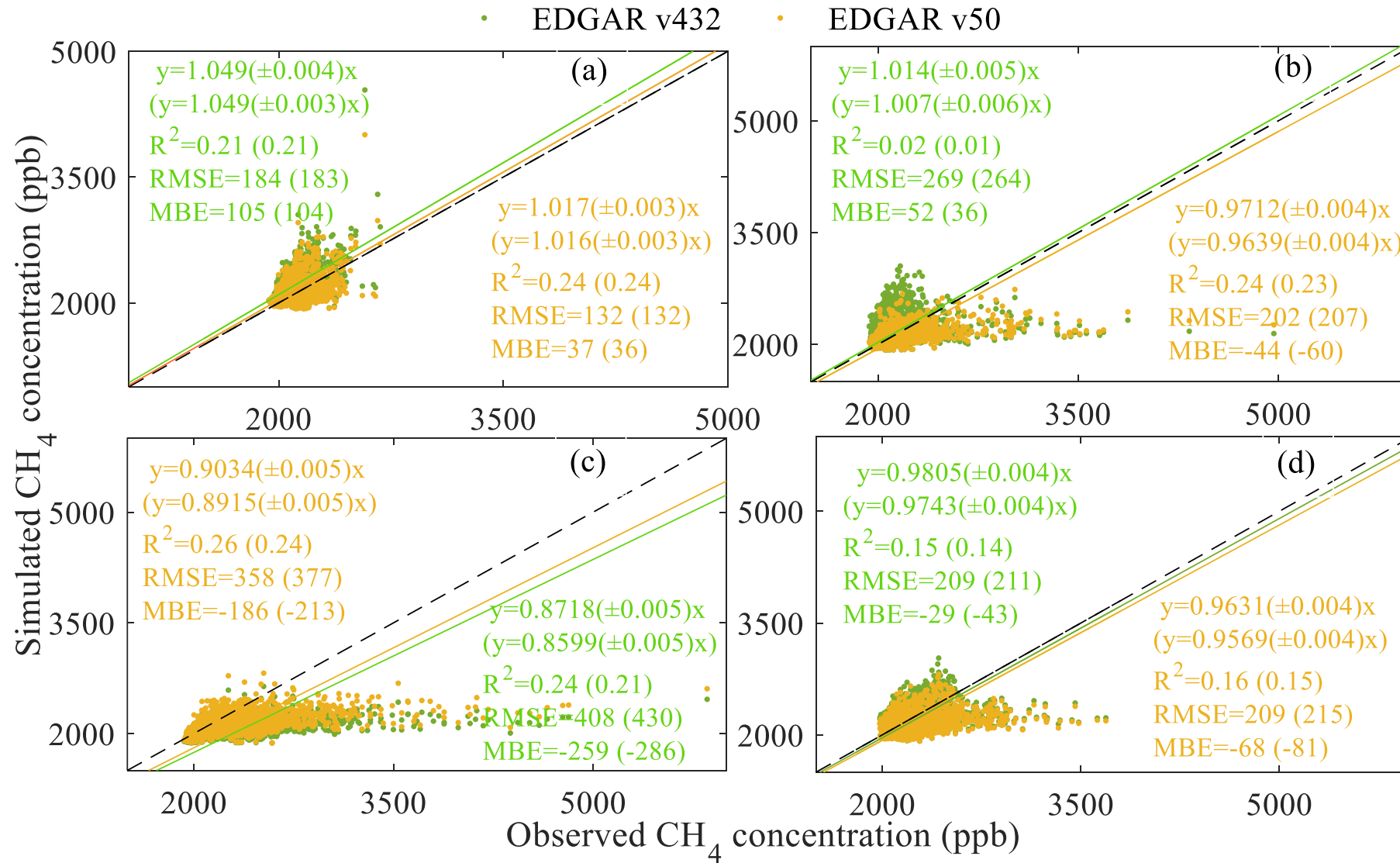
(a) Winter
(b) Spring
(c) Summer
(d) Autumn

3 Results & discussion



3 Results & discussion

Including emissions from wetlands



3 Results & discussion

Scaling factors for the main methane sources on a monthly time scale based on EDGAR v50

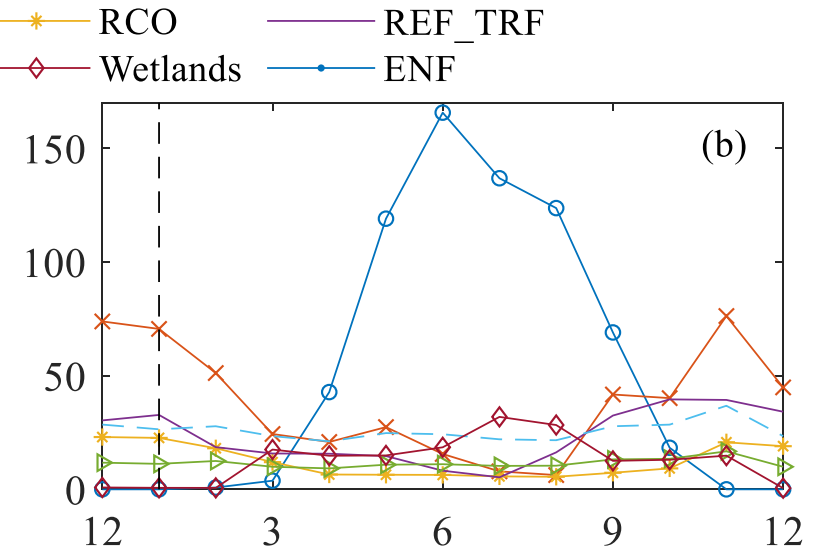
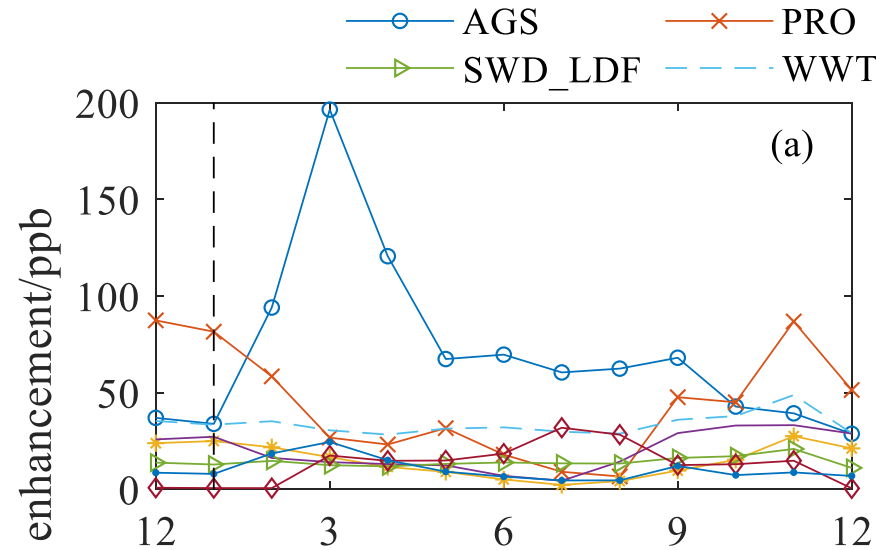
Time	AGS	PRO	RCO	REF_TRF	SWD_LDF	WWT	Wetlands	Other
Dec 2017	1.36	0.47	1.05	0.02	1.36	1.07	1.36	1.36
Jan 2018	1.36	0.24	1.22	0.45	1.36	1.20	1.36	1.36
Feb 2018	0.93	0.39	0.78	0.01	0.93	0.45	0.93	0.93
Mar 2018	1.32	0.10	0.93	0.14	1.32	0.77	1.32	1.32
Apr 2018	1.75	1.29	1.33	0.15	1.33	0.90	1.33	1.33
May 2018	1.55	1.75	1.11	0.72	1.11	0.76	1.11	1.11
Jun 2018	1.56	1.57	1.54	1.54	1.54	1.00	0.82	1.54
Jul 2018	2.22	1.31	1.31	1.31	1.16	1.30	1.26	1.31
Aug 2018	2.15	1.01	1.01	0.82	1.01	0.92	1.61	1.01
Sept 2018	2.66	1.40	1.48	0.61	1.48	0.90	1.48	1.48
Oct 2018	1.64	0.73	1.50	0.20	1.50	0.76	1.50	1.50
Nov 2018	1.38	0.67	1.05	0.31	1.38	1.07	1.38	1.38
Dec 2018	1.74	0.39	1.08	0.14	1.74	0.90	1.74	1.74

AGS = Agricultural soil; PRO = fuel exploitation; RCO = Energy for building; REF_TRF = Oil refineries and Transformation industry; SWD_LDF = Solid waste landfills; WWT = Waste water handling; Wetlands = wetlands (Marshland and mudflats) and water bodies (rivers, lakes, ponds and reservoirs); Other = the sum of remaining categories (depending on the month). The space filled with gray indicates that the emission source was classified as other in a specific month.

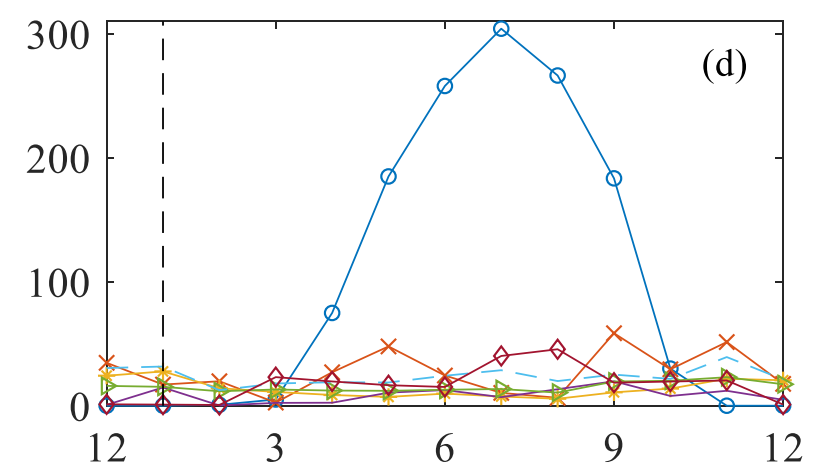
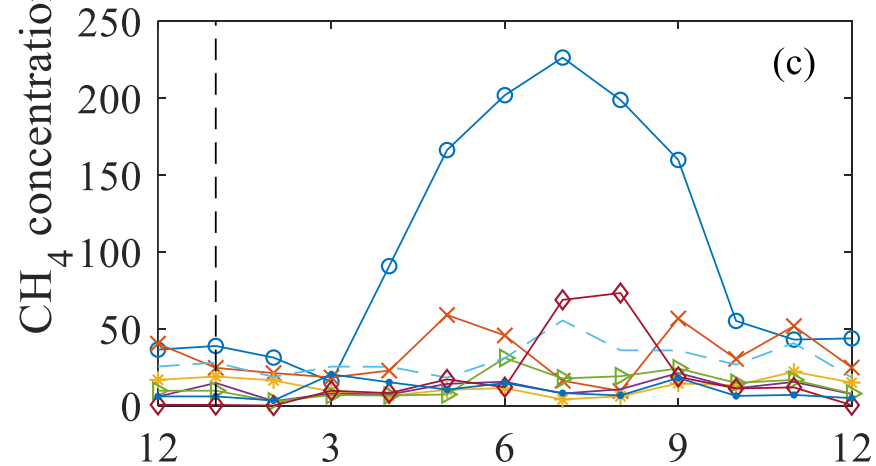
EDGAR v432

EDGAR v50

Original
simulation

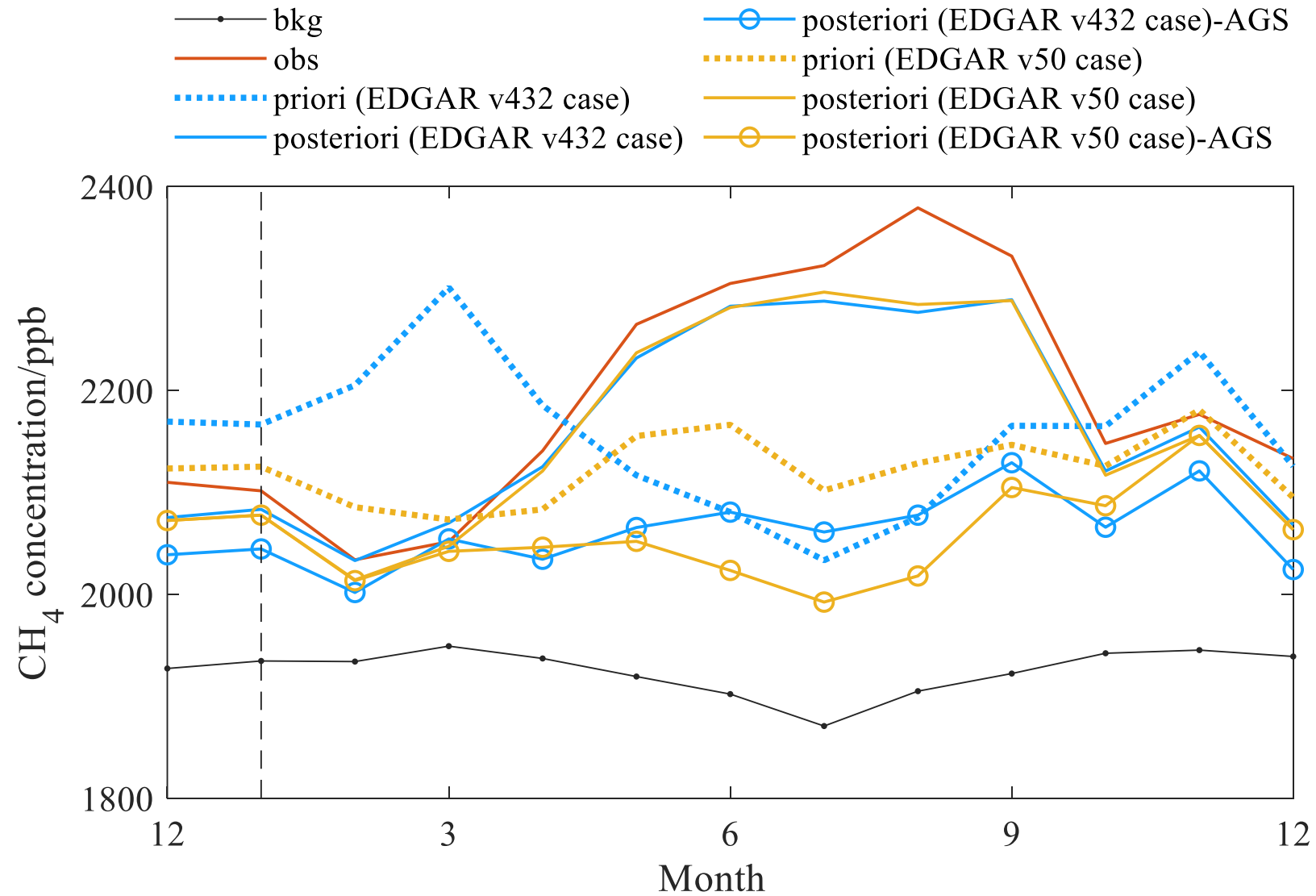


After using
SFBI method



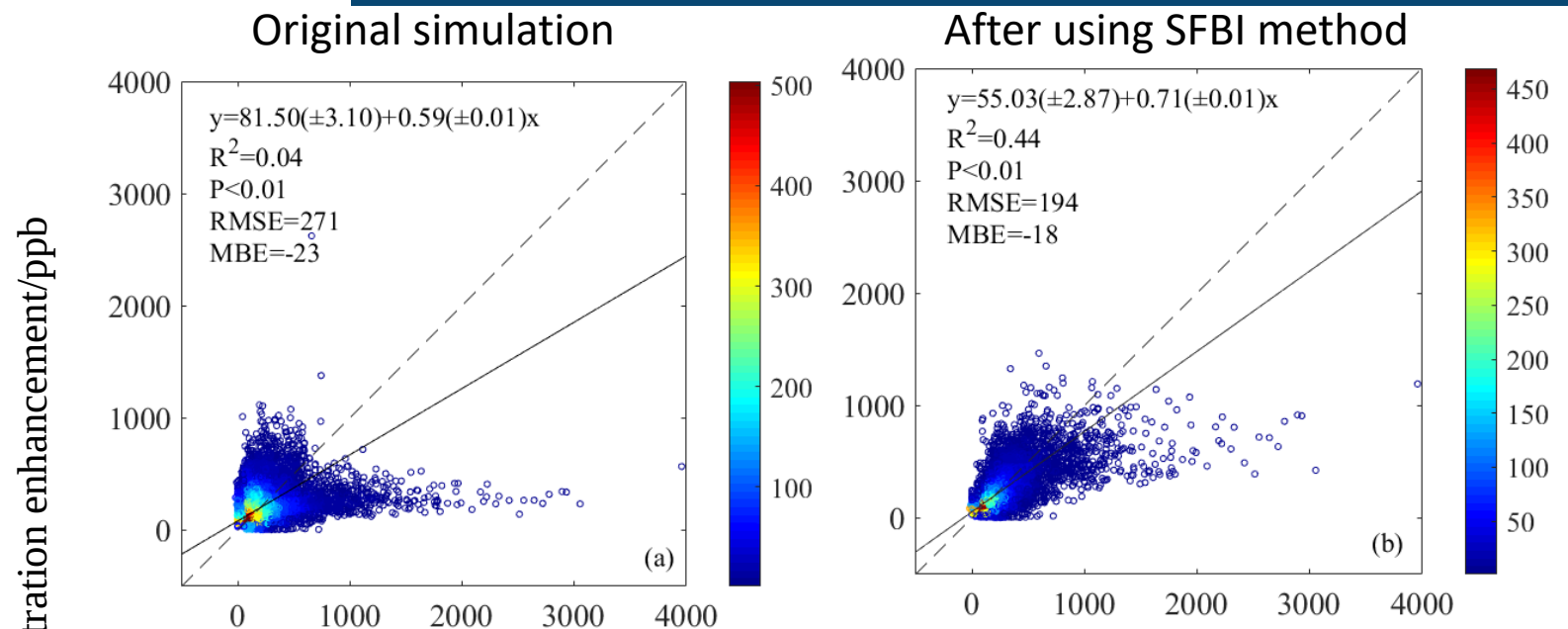
Month

3 Results & discussion

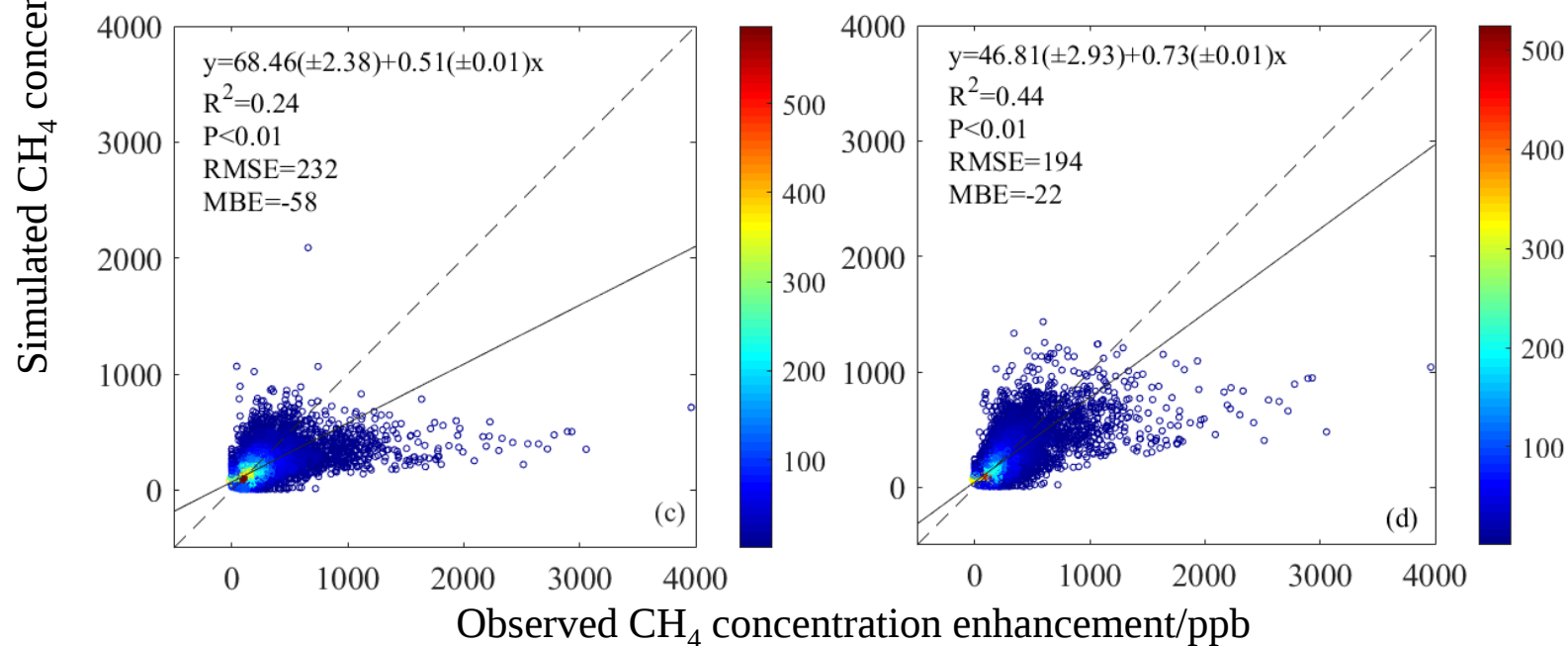


3 Results & discussion

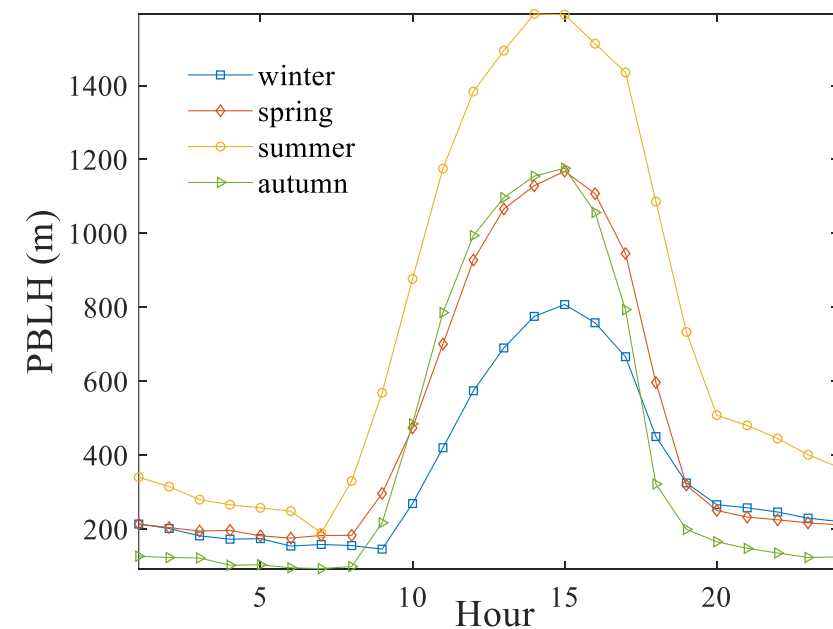
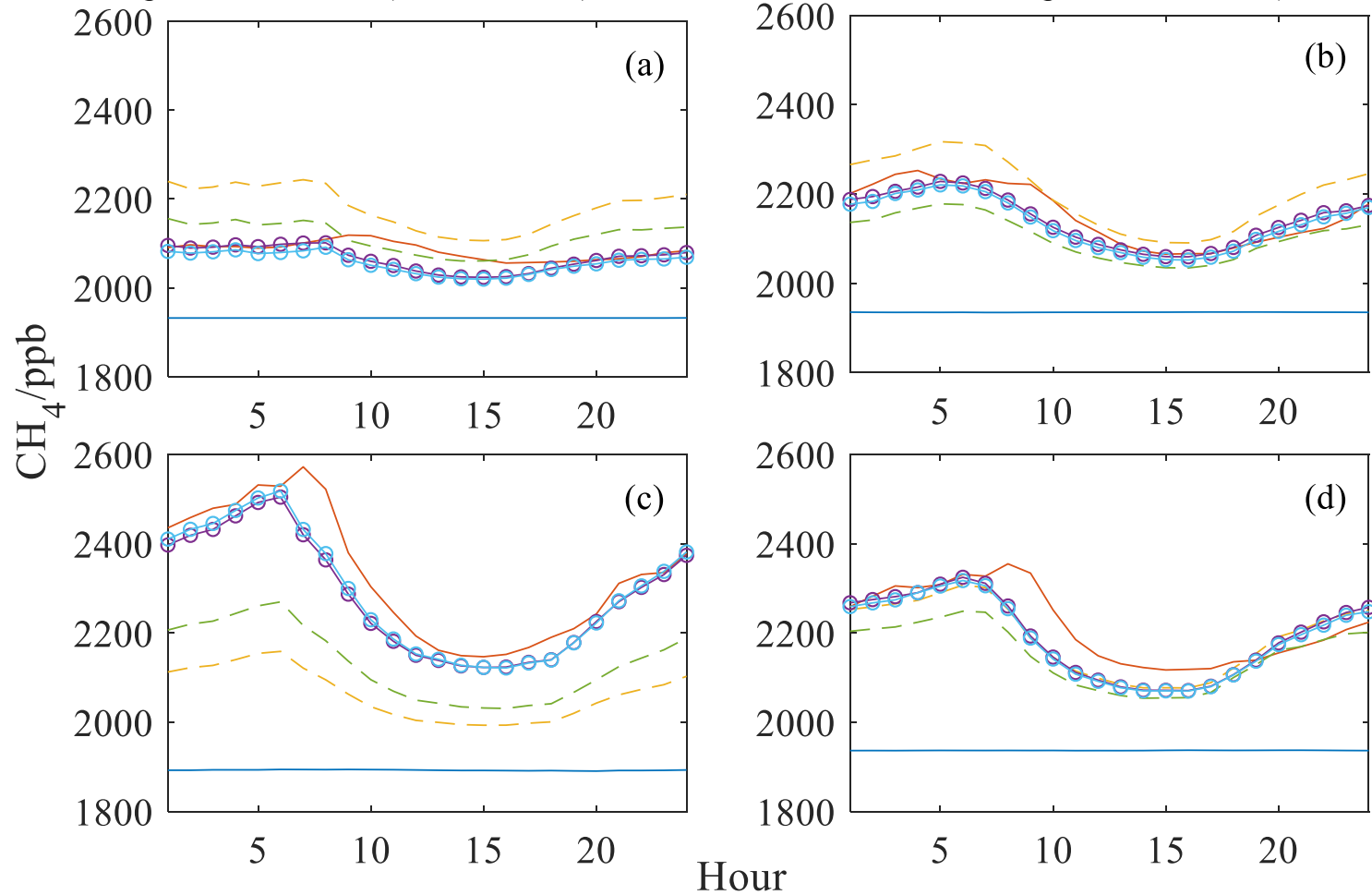
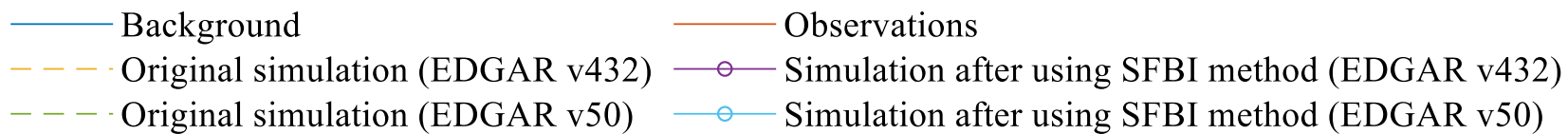
EDGAR v432



EDGAR v50



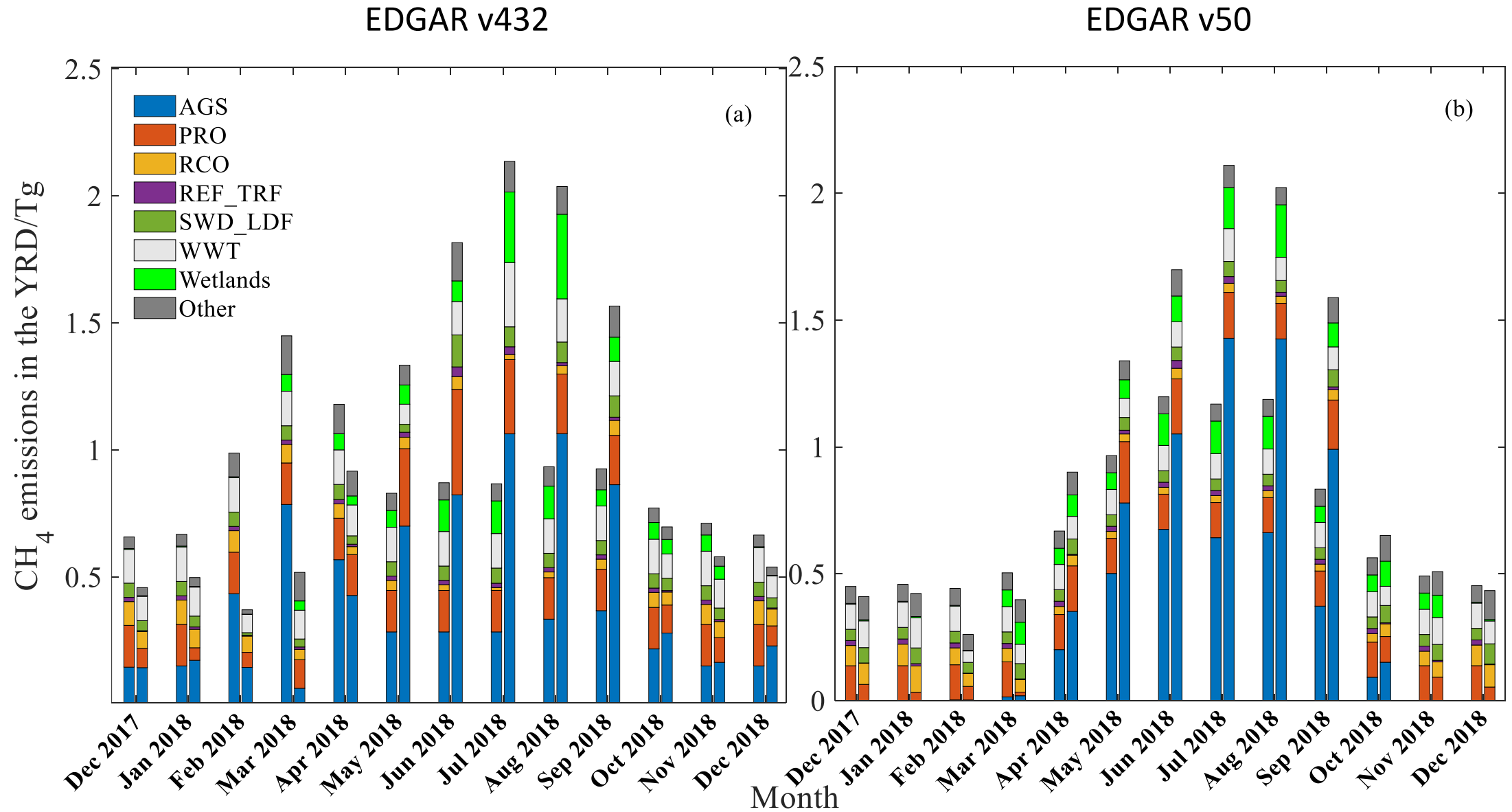
3 Results & discussion



(a) Winter
(b) Spring
(c) Summer
(d) Autumn

Date	EDGAR v50	AGS	PRO	RCO	REF_ TRF	SWD_ LDF	WWT	Wetlands	Other	All
Dec 2017	<i>a priori</i>	0.00	8.96	5.18	1.26	2.87	6.40	0.21	4.32	29.20
	<i>posteriori</i>	0.00	4.20	5.45	0.03	3.91	6.83	0.29	5.89	26.60
Jan 2018	<i>a priori</i>	0.00	8.97	5.50	1.32	2.96	6.45	0.21	4.35	29.76
	<i>posteriori</i>	0.00	2.17	6.72	0.60	4.02	7.72	0.29	5.91	27.42
Feb 2018	<i>a priori</i>	0.26	9.93	4.77	1.40	3.27	7.14	0.21	4.79	31.78
	<i>posteriori</i>	0.24	3.83	3.72	0.01	3.06	3.22	0.20	4.47	18.77
Mar 2018	<i>a priori</i>	0.99	8.97	3.45	1.24	2.96	6.45	4.27	4.36	32.69
	<i>posteriori</i>	1.30	0.93	3.20	0.18	3.90	4.94	5.63	5.75	25.83
Apr 2018	<i>a priori</i>	13.50	9.27	2.14	1.37	3.06	6.66	4.27	4.53	44.79
	<i>posteriori</i>	23.61	12.00	2.83	0.21	4.05	5.97	5.66	6.01	60.35
May 2018	<i>a priori</i>	32.53	8.97	1.76	1.30	2.96	6.45	4.27	4.37	62.61
	<i>posteriori</i>	50.53	15.67	1.95	0.94	3.28	4.88	4.74	4.85	86.83
Jun 2018	<i>a priori</i>	45.25	9.27	1.82	1.34	3.06	6.66	8.33	4.50	80.23
	<i>posteriori</i>	70.43	14.52	2.80	2.06	3.56	6.64	6.81	6.92	113.74
Jul 2018	<i>a priori</i>	41.66	8.97	1.76	1.31	2.96	6.45	8.33	4.34	75.78
	<i>posteriori</i>	92.58	11.73	2.31	1.72	3.87	8.39	10.49	5.67	136.75
Aug 2018	<i>a priori</i>	42.92	8.97	1.76	1.25	2.96	6.45	8.33	4.33	76.97
	<i>posteriori</i>	92.41	9.10	1.79	1.02	3.00	5.93	13.41	4.40	131.07
Sept 2018	<i>a priori</i>	24.98	9.27	1.82	1.28	3.06	6.66	4.27	4.50	55.84
	<i>posteriori</i>	66.38	12.99	2.70	0.78	4.53	6.03	6.33	6.67	106.40
Oct 2018	<i>a priori</i>	6.02	8.97	2.16	1.32	2.96	6.45	4.27	4.39	36.53
	<i>posteriori</i>	9.85	6.57	3.24	0.26	4.43	4.90	6.40	6.57	42.23
Nov 2018	<i>a priori</i>	0.00	9.27	3.83	1.35	3.06	6.66	4.27	4.53	32.96
	<i>posteriori</i>	0.00	6.24	4.02	0.41	4.21	7.11	5.88	6.23	34.10
Dec 2018	<i>a priori</i>	0.00	8.97	5.27	1.31	2.96	6.45	0.21	4.24	29.40
	<i>posteriori</i>	0.00	3.52	5.70	0.18	5.16	5.83	0.37	7.39	28.15

unit: nmol/m²/s
(EDGARv50).



CH₄ emissions in the YRD in 2018

		Emissions from agricultural soils (Tg)	Anthropogenic emissions (Tg)	Emissions from wetlands (Tg)	Total (Tg)
EDGAR v432	<i>a priori</i>	4.01±2.98	10.08±3.51	0.78±0.62	10.87±3.56
	<i>posteriori</i>	6.00±2.99	11.97±3.39	1.05±0.91	13.01±3.51
EDGAR v50	<i>a priori</i>	3.17±2.99	8.16±3.36	0.78±0.62	8.94±3.41
	<i>posteriori</i>	6.21±2.61	11.34±2.92	1.01±0.70	12.35±3.00

Agricultural statistics show that 90% of the rice grown in the YRD is medium and late-season rice (China Statistics Bureau, 2019) and the length of growing season was assumed to be 120 d (17th Jun- 14th Oct).

Emissions during the growing season accounted for about 58.4% for EDGAR v432 and 71.0% for EDGAR v50 of the *posteriori* CH₄ emissions from AGS, respectively.



4

Conclusions

- ❑ The main factor causing [seasonal changes](#) in atmospheric CH₄ concentration in the YRD was [rice cultivation](#).
- ❑ The *posteriori* anthropogenic emissions were [11.97 Tg](#) and [11.33 Tg](#) using EDGAR v432 and EDGAR v50 and the proportion of AGS ranged from 50.1% to 54.8% respectively. The *posteriori* emissions from natural sources (including wetlands and water bodies) were [1.05 Tg](#) and [1.01 Tg](#) in YRD in 2018.
- ❑ The underestimation of the anthropogenic CH₄ emissions in the YRD from the inventory products was mainly due to the [underestimation](#) of emissions from [AGS](#), especially [in growing seasons](#). If this emission source was not considered, the deviation between the estimated value of other anthropogenic emission sources and the *posteriori* results was small (0.10 Tg in EDGAR v432 and 0.14 Tg in EDGAR v50).



T h a n k s

